## CLAIMS

	1.	Data classification apparatus comprising:
5		an input device for receiving a plurality of training
		classified examples and at least one unclassified
		example;
		a memory for storing the classified and unclassified
		examples;
10		an output terminal for outputting a predicted
		classification for the at least one unclassified example;
		and
		a processor for identifying the predicted classification of
		the at least one unclassified example
15		wherein the processor includes:
		classification allocation means for allocating potential
		classifications to each unclassified example and for
		generating a plurality of classification sets, each
		classification set containing the plurality of training
20		classified examples and the at least one unclassified
		example with its allocated potential classification;
		assay means for determining a strangeness value valid
		under the iid assumption for each classification set;
		a comparative device for selecting the classification set to
25		which the most likely allocated potential classification for
		the at least one unclassified example belongs, wherein
		the predicted classification output by the output
		terminal is the most likely allocated classification
		according to the strangeness values assigned by the
30		assay means; and

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a strength of prediction monitoring device for determining a confidence value for the predicted classification on the basis of the strangeness value assigned by the assay means to one of the classification sets to which the second most likely allocated potential classification of the at least one unclassified example belongs.

- 2. Data classification apparatus as claimed in claim 1, wherein the processor further includes an example valuation device which determines individual strangeness values for each training classified example and the at least one unclassified example having an allocated potential classification.
  - 3. Data classification apparatus as claimed in claim 2, wherein Lagrange multipliers are used to determine the individual strangeness value.
- 20 4. Data classification apparatus as claimed in claim 2, wherein the assay means determines a strangeness value for each classification set in dependence on the individual strangeness values of each example.
- 25 5. Data classification apparatus comprising:

  an input device for receiving a plurality of training classified examples and at least one unclassified example;

  a memory for storing the classified and unclassified examples;

stored programs including an example classification program;

an output terminal for outputting a predicted classification for the at least one unclassified example;

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a processor controlled by the stored programs for identifying the predicted classification of the at least one unclassified example wherein the processor includes:

classification allocation means for allocating potential classifications to each unclassified example and for generating a plurality of classification sets, each classification set containing the plurality of training classified examples and the at least one unclassified example with its allocated potential classification;

assay means for determining a strangeness value valid under the iid assumption for each classification set;

a comparative device for selecting the classification set to which the most likely allocated potential classification for the at least one unclassified example belongs, wherein the predicted classification output by the output terminal is the most likely allocated potential classification according to the strangeness values assigned by the assay means and

a strength of prediction monitoring device for determining a confidence value for the predicted classification on the basis of the strangeness value assigned by the assay means to one of the classification sets to which the second most likely allocated potential classification of the at least one unclassified example

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A data classification method comprising: 6. inputting a plurality of training classified examples and at least one unclassified example; identifying a predicted classification of the at least one unclassified example which includes, 5 allocating potential classifications to each unclassified example; generating \a plurality of classification sets, each classification\set containing the plurality of training classified examples and the at least one unclassified 10 example with its allocated potential classification; determining a strangeness value valid under the iid assumption for each classification set; selecting the classification set to which the most likely allocated potential classification for the at least one 15 unclassified example belongs, wherein the predicted classification is the most likely allocated potential classification in dependence on the strangeness values; determining a confidence \ value for the predicted classification on the basis of the strangeness value 20 assigned to one of the classification sets to which the

second most likely allocated potential classification for the at least one unclassified example belongs; and outputting the predicted classification for the at least one unclassified example and the confidence value for the predicted classification.

7. A data classification method as claimed in claim 6, further including determining individual strangeness values for each training classified example and the at

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least one unclassified example having an allocated potential classification.

- 8. A data classification method as claimed in any one of the preceding claims, wherein the selected classification set is selected without the application of any general rules determined from the training set.
- 9. A data carrier on which is stored a classification program 10 for classifying data by performing the following steps: generating a plurality of classification sets, each classification set containing a plurality of training classified examples and At least one unclassified example that has been allocated a potential classification;
- 15 determining a strangeriess value valid under the iid assumption for each/classification set;
  - selecting the classification set to which the most likely allocated potential classification for the at least one unclassified example belongs, wherein the predicted classification is the most likely allocated potential classification in dependence on the strangeness values; and

value for the predicted determining a confidence classification on the basis of the strangeness value assigned to one of the classification sets to which the second most likely allocated potential classification for the at least one unclassified example belongs.

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